



## Linear Transformations

Much of linear algebra can be understood as the study of *linear transformations*. Here, *transformation* is just another fancy word for function. While *linear* just specifies that this function satisfies a few useful properties.

More precisely, we say that a function  $T : \mathbb{R}^m \rightarrow \mathbb{R}^n$  is a *linear transformation* if it satisfies the two following properties for any scalar  $c$  and any vectors  $\mathbf{u}$  and  $\mathbf{v}$ :

1.  $T(c\mathbf{v}) = cT(\mathbf{v})$
2.  $T(\mathbf{u} + \mathbf{v}) = T(\mathbf{u}) + T(\mathbf{v})$

Many common transformations turn out to be linear: reflections, rotations, contraction, dilation ...

## Finding the Matrix of a Linear Transformation

One really useful fact about linear transformations is that every linear transformation can be represented as a matrix.

To see how, let's consider an  $m$ -dimensional vector  $\mathbf{v}$ , which we can write as

$$\mathbf{v} = \begin{bmatrix} v_1 \\ v_2 \\ v_3 \\ \vdots \\ v_m \end{bmatrix}$$

We can split this vector up in the following way:

$$\mathbf{v} = v_1 \begin{bmatrix} 1 \\ 0 \\ 0 \\ \vdots \\ 0 \end{bmatrix} + v_2 \begin{bmatrix} 0 \\ 1 \\ 0 \\ \vdots \\ 0 \end{bmatrix} + v_3 \begin{bmatrix} 0 \\ 0 \\ 1 \\ \vdots \\ 0 \end{bmatrix} + \dots + v_m \begin{bmatrix} 0 \\ 0 \\ 0 \\ \vdots \\ 1 \end{bmatrix}$$

Each of these vectors with a 1 in the  $i$ -th entry and a 0 everywhere else is known as a *standard basis vector*, and is usually written as  $\mathbf{e}_i$ . So we can also write the above in shorthand as:

$$\mathbf{v} = v_1\mathbf{e}_1 + v_2\mathbf{e}_2 + v_3\mathbf{e}_3 + \dots + v_m\mathbf{e}_m$$

Now, suppose we have a linear transformation  $T : \mathbb{R}^m \rightarrow \mathbb{R}^n$ , and we want to figure out what  $T(\mathbf{v})$  is. Then,

$$T(\mathbf{v}) = T(v_1\mathbf{e}_1 + v_2\mathbf{e}_2 + v_3\mathbf{e}_3 + \dots + v_m\mathbf{e}_m)$$

Using the second property of linear transformations,

$$T(\mathbf{v}) = T(v_1\mathbf{e}_1) + T(v_2\mathbf{e}_2) + T(v_3\mathbf{e}_3) + \dots + T(v_m\mathbf{e}_m)$$

Then using the first property of linear transformations,

$$T(\mathbf{v}) = v_1T(\mathbf{e}_1) + v_2T(\mathbf{e}_2) + v_3T(\mathbf{e}_3) + \dots + v_mT(\mathbf{e}_m)$$

It turns that we can write the right hand side as a matrix-vector product, which looks like:

$$T(\mathbf{v}) = \begin{bmatrix} T(\mathbf{e}_1) & T(\mathbf{e}_2) & T(\mathbf{e}_3) & \dots & T(\mathbf{e}_m) \end{bmatrix} \cdot \begin{bmatrix} v_1 \\ v_2 \\ v_3 \\ \vdots \\ v_m \end{bmatrix}$$

$$T(\mathbf{v}) = \begin{bmatrix} T(\mathbf{e}_1) & T(\mathbf{e}_2) & T(\mathbf{e}_3) & \dots & T(\mathbf{e}_m) \end{bmatrix} \cdot \mathbf{v}$$

where the  $i$ -th column of the matrix is given by the  $n$ -dimensional vector  $T(\mathbf{e}_i)$ .

So given any linear transformation  $T$ , we can find its matrix by calculating each  $T(\mathbf{e}_i)$  and using it as the  $i$ -th column.